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# Neural circuit function redundancy in brain disorders

Beatriz E. P. Mizusaki and Cian O'Donnell

## Abstract

Redundancy is a ubiquitous property of the nervous system. This means that vastly different configurations of cellular and synaptic components can enable the same neural circuit functions. However, until recently, very little brain disorder research has considered the implications of this characteristic when designing experiments or interpreting data. Here, we first summarise the evidence for redundancy in healthy brains, explaining redundancy and three related sub-concepts: sloppiness, dependencies and multiple solutions. We then lay out key implications for brain disorder research, covering recent examples of redundancy effects in experimental studies on psychiatric disorders. Finally, we give predictions for future experiments based on these concepts.

## Addresses

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## Ubiquity of redundancy in the nervous system

Neural circuits have an astronomically large space of potential configurations of their molecular, cellular and synaptic components. Somehow these components must be arranged to enable the circuit to perform useful computations. The task is made easier by the ubiquitous phenomenon of redundancy, which is the idea that, within this enormous space of all possible cellular component configurations, there exists a large subset that achieves effectively equivalent macroscopic computations [1,2]. The main empirical evidence for redundancy in neural systems comes from a series of classic studies from Eve Marder [1,3] on a small neural circuit from the crab and lobster stomatogastric ganglia (STG). Using

computational models, they found that very different arrangements of each STG neuron's ion channels and synaptic conductances could achieve identically sequenced circuit oscillations [4,5]. Accordingly, in experiments, these neurons showed twofold to threefold heterogeneity in cellular properties across animals, despite exhibiting consistent circuit function [6]. Similar redundancy phenomena have also been described in Hodgkin–Huxley models [7], mammalian pyramidal neuron models [8], tadpole neurons [9], rodent neuronal activity *in vitro* and *in vivo* [10,11] and human neuroimaging data [12]. Collectively, these studies, plus theoretical arguments [2,13–15], suggest that redundancy is a universal property of the nervous system.

In addition to the core idea of redundancy, we describe three further sub-concepts: sloppiness, compensation and multiple solutions. Sloppiness is the idea that high-level circuit properties are not equally sensitive to the properties of each of its components. Perturbations to some of these components may result in extreme changes to overall function, whereas others may even vary widely while incurring little effect at the circuit level. Dependence is a developmental phenomenon where multiple circuit parts are co-tuned with each other, with strong dependencies between their effects on overall function. We consider multiple solutions as the observation that the various configurations of cellular components that enable satisfactory circuit-level functions need not be connected with each other: multiple functional islands can co-exist in the parameter space.

Despite the ubiquity of redundancy in the brain, surprisingly little research on brain disorders has considered its implications when designing experiments or interpreting data. In the remainder of this review, we will elaborate these implications and outline how they can be used to guide future brain disorder research.

## Implications of neural redundancy for brain disorder research

The phenomenon of redundancy and each of its three sub-phenomena (sloppiness, dependencies and multiple solutions) have distinct implications for brain disorders (Figure 1). First, we illustrate the effects of redundancy itself (Figure 1a) through a measure of the performance of a hypothetical circuit's function shown in a contour plot relative to component parameters,  $\theta_1$

**Box 1. Convergence of brain dysfunction at the level of neural circuits**

Recent high-powered genetic studies have uncovered myriad mutations that correlate with statistical risk for neurodevelopmental and psychiatric disorders [16]. For example, ~100 distinct genetic mutations have been found that elevate risk for schizophrenia [17], as well as another ~100 that increase risk for autism spectrum disorder (ASD) [18], including overlapping risk gene sets across different psychiatric disorders [19,20]. Despite this heterogeneity at the genetic level, patients may present overlapping symptoms at the cognitive level, and so receive the same umbrella diagnoses. This implies that there must be points of phenotypic convergence within the levels of organisation in the nervous system, which span from molecules to cells, circuits, cognition and behaviour. Neural circuits are a promising focus for analysis for two reasons: first, if molecular-, synaptic- or cellular-level alterations in a brain disorder do not lead to alterations in neural circuit function, then they cannot be contributing to cognitive symptoms. Second, because neural circuits are closer to behaviour than cellular components are, circuit-level interventions may have more predictable effects on cognitive symptoms, compared with cellular- or molecular-level interventions. This argues that in a symptom-targeted approach, we should bias our efforts towards understanding, diagnosing and treating brain disorders at the neural circuit level rather than the cellular or molecular level, as is common in drug development today [21–24].

and  $\theta_2$ . Darker shades of pink correspond to better performance of the circuit. Real systems actually contain thousands of key components, so the parameter space would be much higher-dimensional: our two-dimensional plot is an oversimplification to aid visualisation. Because of evolutionary pressure, we can assume measurements from wild-type animals or neurotypical people will be located near the peak (blue circles) [25]. As an example, two different genetic mutations linked to the same brain disorder may lead to changes in both parameters, drifting affected individuals to different points in the parameter space. Although each genetic mutation may shift the mean parameter changes in a different direction away from the neurotypical case, redundant disorders end up on roughly the same contour line with respect to circuit function. This implies that from the circuit point of view, these distinct mutations manifest with the same phenotype, even if their parameters differ.

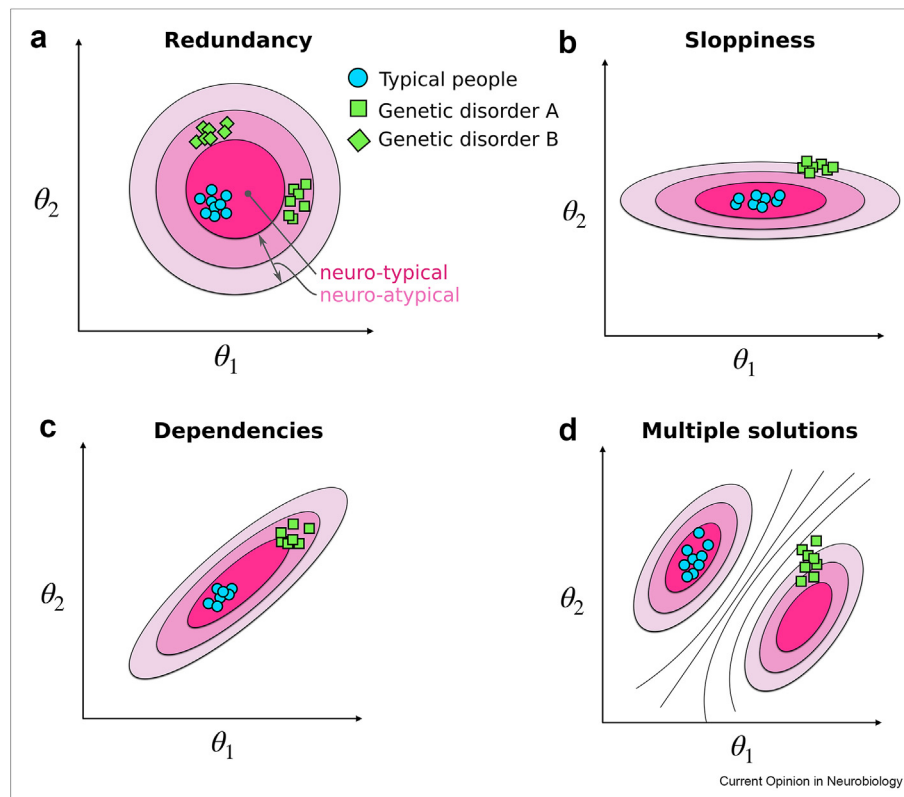
Importantly, however, despite their similarity in circuit function, the two clusters of individuals with brain disorders might be differentially susceptible to perturbations. In the example of the crustacean STG, individual animals may have distinctive sensitivities to changes in temperature, pH or neuromodulators [26–29]. In our example (Figure 1a), we can imagine some environmental effector such as a drug or stressful life event that causes a small increase in  $\theta_1$ , corresponding to a rightward shift in all the data points. For both neurotypical people and those with genetic disorder A, this effect would be benign as it would not cause a change in

circuit function. In contrast, the same effector could push those with mutation B into even worse values. Alternatively, a different effector that increased  $\theta_2$  would not cause a circuit function change in either genetically typical people or those with genetic disorder B, but would have a deleterious effect on those with mutation A. This also illustrates a phenomenon with proposed treatments—they may work to rescue symptoms in one group of patients but not another, even if both groups appear superficially similar. In this sense, redundancy might not only be hiding latent vulnerabilities in the system but also heterogeneities in those vulnerabilities across patient groups.

Second, molecular or cellular alterations observed in tissue from human patients or animal models may not actually be affecting the circuit-level function—they may be benign. The circuit may be robust to changes in these components over some tolerable range. This property is referred to by different names, according to the research field or author. We will refer to it here as sloppiness [30,31]. Within the same schematic as before, sloppiness can be seen on another hypothetical contour plot (Figure 1b). In this case, the circuit function is relatively insensitive to the exact value of one parameter ( $\theta_1$ ), so it may vary horizontally in the plot across a large range without causing much change in circuit function. In contrast, small changes to the other parameter ( $\theta_2$ ) will induce large changes in circuit function. In this case,  $\theta_1$  is the sloppy parameter. If we consider a brain disorder where a genetic mutation tends to increase both parameters  $\theta_1$  and  $\theta_2$  in the brains of affected individuals, the change in  $\theta_2$  would be the primary driver of dysfunction, although  $\theta_1$ 's value would still be correlated with disease severity. If an experimental scientist measured the value of  $\theta_1$  in both wild-type and brain disorder animal models, they may see a clear difference in the group mean values of  $\theta_1$  and a parallel change in circuit function. They may conclude that the changes in  $\theta_1$  are responsible for the circuit-level deficits and design an intervention to reverse the molecular level change in  $\theta_1$ ; however, the treatment would not be successful.

Third, we comment on redundant dependencies: altered components may individually have large effects on circuit function when perturbed genetically or experimentally, but homeostatic processes during development may restore high-level function by compensating with changes in other circuit components. In a simple case, this could be a straight pairing of opposing factors, such as increased expression of sodium channels that depolarise the cell being counteracted by increased expression of potassium channels that hyperpolarize it. However, in intact brains, there are so many nonlinear interactions that the compensatory relationships might not be obvious from raw measurements. In (Figure 1c), we depict this idea with another

Figure 1



**Forms of redundancy.** The performance of a hypothetical neural circuit is shown as a contour map in pink as a function of the values of two of its components,  $\theta_1$  and  $\theta_2$ . Darker hues of pink represent better circuit performance. Symbols show possible measured values of  $\theta_1$  and  $\theta_2$  for genetically typical people (blue circles) and two different genetic brain disorders A and B (green squares and diamonds, respectively). Panels A–D show different versions of the contour map illustrating various forms of redundancy: (a) generic redundancy, (b) sloppiness, (c) dependencies and (d) multiple solutions.

hypothetical contour map on a two-dimensional parameter space. In this case, proper circuit function requires jointly low or jointly high values of  $\theta_1$  and  $\theta_2$  together, so if one parameter is low while the other is high, then circuit function is impaired. A genetic mutation could cause a direct increase in  $\theta_1$ , but be developmentally compensated by a corresponding increase in  $\theta_2$ . In this situation, an experiment may yield clear group-level differences in  $\theta_1$  between wild-type and brain-disorder animal models, but they may not measure parallel changes in  $\theta_2$ . If a scientist nevertheless found a behavioural phenotype due to unobserved alterations elsewhere in the brain, they might go on to design an intervention to bring the value of  $\theta_1$  in the animal model back down to wild-type values without altering  $\theta_2$ , which might inadvertently make the circuit function worse, not better.

Last, there may be multiple distinct optima to the circuit design (Figure 1d), appearing as multiple islands. Although it is likely that these peaks may be connected via some paths in the full high-dimensional parameter space of all circuit components [32,33], any

experimental measurement of a small subset of parameters or therapeutic intervention may have access to only a low-dimensional subspace, where such local optima are likely to persist. Even though the phenomenon of multiple solutions complicates our attempts to understand how brains work, it could paradoxically end up simplifying our search for brain disorder interventions. It implies that fixing circuit function does not require a direct reversal of the original alteration. Depending on how many solutions exist, it may instead be more practical to find a new configuration that restores the circuit operating mode, rather than trying to undo all the various component changes that have accumulated across development—most of which are in any case likely to be hidden to the experimentalist or clinician. We illustrate the phenomenon of multiple solutions schematically in (Figure 1d). The nearest part of parameter space that rescues circuit function in the brain disorder case is not the same as for the genetically typical case. In addition, the intervention that implements this correction would involve changing only  $\theta_2$ , even though  $\theta_1$  was the parameter altered by the original genetic mutation. Therefore, the phenomenon of

multiple solutions may open up counterintuitive options for therapeutics.

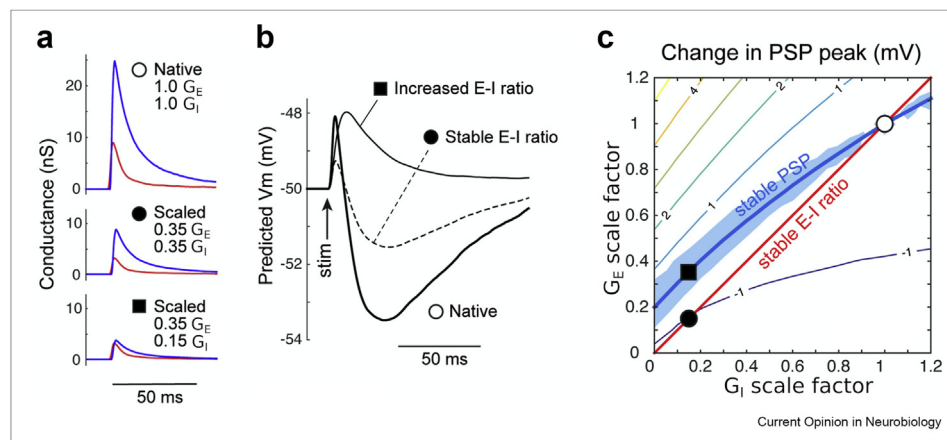
### Empirical examples of redundancy in brain disorders

Although few studies have directly explored the consequences of redundancy in brain disorders, many have found evidence for homeostatic compensation where changes in one brain component seemed to be counterbalanced by changes in others, a form of redundancy [34–39]. There are also evidence for disrupted homeostatic plasticity [40–42] and proposals for how global brain perturbations could lead to deficits only in select neural circuits [43,44]. However, one recent study by Antoine et al. [45] found explicit evidence for circuit redundancy in mouse models of autism (Figure 2). The authors used patch-clamp electrophysiology to measure excitatory and inhibitory synaptic inputs from layer 4 onto single-layer 2/3 pyramidal neurons in brain slices of the primary somatosensory cortex from wild-type mice and from four different genetic mouse models of autism. Nominally, their aim was to ask if the ratio of synaptic excitation to inhibition (E/I balance) was altered in the autism mouse models, a common theory for autism [46,47]. Indeed, they found that in each of the four autism models, inhibition was decreased more than excitation, implying an increase in the E/I ratio compared with wild-type mice, but surprisingly, they also found that in each case, the amplitude of postsynaptic potentials (PSPs) and spiking responses to stimulation was

unchanged relative to wild-type controls. The authors explained this mismatch via computational modelling, which showed that a range of different synaptic E/I ratios would be consistent with any given PSP amplitude (Figure 2). The contour plot in (Figure 2c) shows the PSP amplitude as a function of excitatory (y-axis) and inhibitory (x-axis) synaptic strengths. The mean wild-type values are marked by the open circle, and the blue curve shows the region in this two-dimensional parameter space where wild-type PSP amplitude is preserved, analogous to the dark pink regions in the plots in (Figure 1). Results from all four autism mouse models sat along the blue curve, with the *Cntnap2* KO values shown in (Figure 2c) as the black square. Although the autism-related genetic mutations were causing real shifts in synaptic properties, their net effects were redundant, causing no change in the neuron's response to synaptic stimulation. Overall, the result suggested that the autism field's decade-long search for E/I imbalance may have been misguided because redundancy nullified its apparent effect on circuit function.

Another recent study, by O'Donnell et al. [48], found using a computational model of the same brain region, the mouse L2/3 somatosensory cortex, that circuit-level function shows extreme differences in sensitivity to perturbations in some components over others, corresponding to sloppiness (Figure 1b). In line with previous studies [49], the authors also found that neural correlations were altered in a mouse model of fragile-X syndrome, but this

Figure 2



**Redundancy in mouse models of autism.** **a:** Example excitatory (red) and inhibitory (blue) synaptic conductance time series from a basic computational model of pyramidal cell voltage. The top 'native' plot shows the case when synaptic conductances are set to the values estimated from layer 4 to layer 2/3 synapses in wild-type mice. In *Cntnap2* knockout animals, a model for autism, excitatory and inhibitory synaptic conductances ( $G_E$  and  $G_I$ ) were decreased to 35% and 15% of wild-type values, respectively, implying an increase in the excitation/inhibition ratio. The middle plot shows traces of both conductances were scaled equally to 35% of wild-type values; the bottom plot shows situation that matches the data, where inhibition is decreased more than excitation. **b:** Compound postsynaptic potentials (PSPs) corresponding to the three scenarios shown in panel A. Note that the PSP amplitude is decreased relative to the native case if the E-I ratio is kept fixed, whereas the increased E-I ratio keeps the PSP amplitude matched to native. **c:** The contour map of the peak PSP amplitude as a function of the scaling factor on excitatory and inhibitory synaptic strengths. The open circle is the mean value from wild-type control animals. The red line corresponds to the fixed E-I ratio; the blue line corresponds to the fixed PSP peak. The black square symbol is the mean value of synaptic strengths in *Cntnap2* knockout mice, whereas the black circle symbol is where values would lie if the E-I ratio was stable. The figure was adapted with permission from Ref. [45].



circuit-function—level property did not map neatly onto any one distinct circuit model component, implying both redundancies and dependencies between parameters (as in (Figure 1c). Together, these two examples of redundancy illustrate the crucial importance of considering the functional properties of neural circuits when interpreting the results of experiments measuring circuit component changes in brain disorders.

## Conclusion and outlook

In summary, because redundancy appears to be a ubiquitous feature of the nervous system, we argue that it should be highlighted when trying to understand or develop treatments for brain disorders. How should these concepts be applied at a practical level to enhance treatment prospects? One general prescription is to aim to simultaneously measure as many circuit components in the same individual as possible, to discover their joint effects on circuit function. However, given the enormous number of candidate neural components to measure, and the complexity of the mapping between circuit components and circuit function, this is currently a challenge even in animal models of brain disorders, never mind individual human patients. Although recording technologies will undoubtedly improve over time, there are no easy solutions to these immense technical obstacles. A second practical problem is that if interventions to tackle selected circuit-function symptoms are designed based on redundancy principles, they may risk knock-on effects on other aspects of circuit function. In general, these effects may be hard to predict *a priori*, but nonclinical neurobiology considerations and quantitative computer modelling simulations may be used to pre-screen treatments and narrow the empirical search space. Despite these challenges, we argue that it is better to acknowledge redundancy phenomena early and factor them into our research programmes and experimental designs, rather than running the risk of wasting time, funding and chasing flawed hypotheses that could be later undermined by redundancy.

On a more positive note, we also believe redundancy offers hope because it means that we may not need to classify and measure every last detail of every form of disorder to develop effective treatments for symptoms. It may turn out that there are generic principles of neural circuit dysfunction that allow us to generalise our insights across the ever-growing list of molecularly distinct brain disorders. These principles may in turn allow us to derive rational treatment strategies that enable correction of common, systems-level symptoms, rather than painstakingly attempting to correct each molecular-level perturbation one at a time.

As this is a very general framework, we anticipate that many predictions follow. We end by giving one example prediction for redundancy and each of its three sub-phenomena.

- Redundancy itself predicts that the magnitude of the differences in measures of neural circuit components between genotypes is greater than the magnitude of differences of measures of functional activity in the same circuits. However, this superficial similarity may hide heterogeneity in response to perturbations, across groups of related disorders.
- Sloppiness predicts that the degree of within- or across-animal heterogeneity in a circuit component parameter should be inversely proportional to the magnitude of its effect on circuit function. If a particular component shows low heterogeneity across wild-type animals, and it is altered in a brain disorder, then it likely also plays a causal role in any circuit-function—level alterations.
- Dependence predicts that any set of cellular components that strongly co-vary within wild-type animals are unlikely to be causally contributing to circuit-function—level alterations in brain disorders. Reversing the changes of any subset of these components in isolation might even exacerbate circuit-function symptoms.
- Multiple solutions predict that the individual animals from a genetically modified cohort that are most similar to the wild-type animals at the circuit function or behavioural level will not necessarily have the most wild-type—like circuit components.

### Box 2. Insights from deep learning

The phenomena of redundancy and multiple optima have also been extensively explored in the field of deep learning and artificial neural networks, where large brain-inspired models are trained to perform computational tasks by iteratively tuning the weights of connections between units, analogous to the synaptic strengths of neural connections in the brain. Deep learning researchers have a key advantage compared with neuroscientists: they can mathematically calculate an unambiguous measure of task performance, unlike in neurobiology where a circuit's performance quality may usually only be guessed. As a result, researchers have explored the actual shapes of deep neural network parameter optimality landscapes in some detail [50]. Their two main relevant findings are as follows: first, these systems tend to have many local optima, but in high-dimensional parameter space, the optima are almost always connected by continuous paths along some small subset of dimensions [33]. The corresponding implication for neuroscience is that there may be some small, special combinations of neural circuit parameters that can be targeted for interventions, which can effectively move the system towards optima while minimising the risk of severe detriments. Second, some optima are wide while others are narrow, with parameters in wide optima resulting in better generalisation performance for new input signals [51,52]. In neuroscience, wide optima may also be desirable from a robustness point of view: they would be more tolerant to biological noise or drift in circuit parameters over time. Wide optima could be selectively targeted when designing brain disorder interventions by probing a range of different parameter values near the optimum, analogous to the ways deep neural network training algorithms are modified to bias the search process towards wide optima.

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## Conflict of interest statement

Nothing declared.

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- \* of special interest
- \*\* of outstanding interest

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 This study examined homeostatic plasticity of intrinsic excitability in cultured cortical neurons from both wild-type and *Fmr1* knock-out mice, a model for Fragile-X Syndrome. They first found that neither baseline excitability nor homeostatic excitability changes following activity deprivation were altered in *Fmr1* knock-outs (KO), according to classic measures such as input resistance and current threshold. Interestingly however, the *Fmr1* KO neurons did not change their firing properties in the same way as the wild-type neurons did under activity deprivation. Such spike properties are arguably more relevant to the neuron's input–output function than cellular parameters like input resistance. Therefore this study highlights the importance of focusing on net neuronal circuit function alterations in brain disorders, rather than the cellular parameters themselves.
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